

# PARTIAL DIFFERENTIAL EQUATIONS

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## TYPES

1) • Elliptic Eq.

e.g.  $\nabla^2 u(\vec{x}) = g(\vec{x}) \quad \vec{x} \in \Omega \subset \mathbb{R}^3$

space only

one BC on every boundary

Dirichlet  $u(\vec{x}) = A(\vec{x}), \quad \vec{x} \in \partial\Omega$

von Neuman  $\frac{\partial u}{\partial n}(\vec{x}) = B(\vec{x}) \quad \vec{x} \in \partial\Omega$

asymptotic falloff  $u(\vec{x}) \rightarrow 0, \quad |\vec{x}| \rightarrow \infty$

solve "all at once"

More generally

$$A^{ij}(\vec{x}) \frac{\partial^2 u}{\partial x^i \partial x^j} + F(u, \partial_i u, \vec{x}) = 0,$$

$A^{ij}(\vec{x})$  positive definite

central for constructing initial data for NR

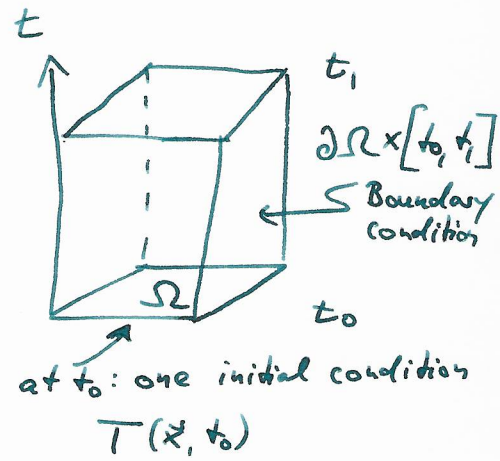
2) Parabolic Eqs

prototype diffusion eq

$$\frac{\partial T(\vec{x}, t)}{\partial t} - \gamma \nabla^2 T(\vec{x}, t) = S(\vec{x}, t)$$

infinite propagation speeds

mostly irrelevant for NR (violate causality),  
except for Apparent Horizon finding



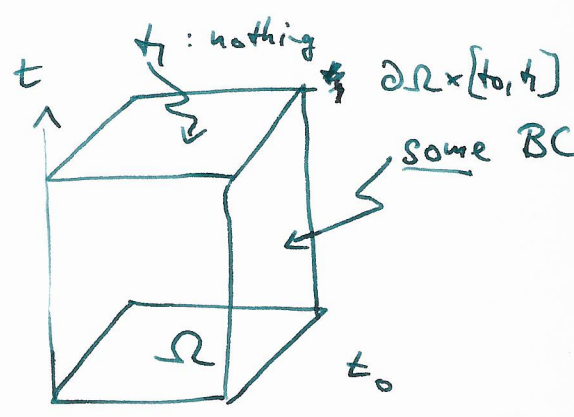
### 3) Hyperbolic PDE

prototype wave-eg

$$\frac{\partial^2 \psi}{\partial t^2} - c^2 \nabla^2 \psi = S$$

$$\psi(\vec{x}, t) \quad \Omega \times \mathbb{R}$$

↑  
time



two initial conditions  
 $\psi(\vec{x}, t_0), \dot{\psi}(\vec{x}, t_0)$

information propagates with finite velocity

⇒ Evolution of GR, hydro, E+M, ...

• First order reduction

define  $\phi_i \equiv c \partial_i \psi \implies c^2 \nabla^2 \psi = c^2 \delta^{ij} \partial_i \partial_j \psi = \delta^{ij} \partial_i \phi_j$

$\pi \equiv -\partial_t \psi \implies \partial_i \pi = -\partial_i \partial_t \psi = -\partial_t \partial_i \psi = -\frac{1}{c} \partial_t \phi_i$

$$\left( \frac{\partial^2}{\partial t^2} - c^2 \nabla^2 \right) \psi = S \iff \begin{cases} \partial_t \psi = -\pi \\ \partial_t \pi + c \delta^{ij} \partial_i \phi_j = -S \\ \partial_t \phi_i + c \partial_i \pi = 0 \end{cases}$$

general 1st order form:  $\underline{u}$  = vector of variables e.g.  $\underline{u} = \{\psi, \pi, \phi_i\}$

$$\| \partial_t \underline{u} + \underline{A}^j \partial_j \underline{u} = \underline{F} \quad (*) \|$$

↑  
may depend on  $\vec{x}, t, \underline{u}$ . But not  $\partial_j \underline{u}$

Examine local properties of solution  $\underline{u}(\vec{x}, t)$  via plane waves with  $k = \frac{2\pi}{\lambda} \rightarrow \infty$  (i.e. short wavelengths)

Ansatz  $\underline{u}(\vec{x}, t) = \underline{u}_0 e^{i(\vec{k}\vec{x} - \omega t)} = \underline{u}_0 e^{ik(\hat{n}\vec{x} - vt)}$   
 $\hat{n}$  = direction of propagation  
 $v = \frac{\omega}{k}$  phase-velocity

in  $\otimes$   $(-ikv \underline{u}_0 + ik \underline{A}_{\hat{n}}^d \underline{u}_0) e^{ik(\hat{n}\vec{x} - vt)} = \underline{F} \quad | : ik$

$$-v \underline{u}_0 + \underline{A}_{\hat{n}}^d \underline{u}_0 = \frac{1}{ik} \underline{F} e^{-ik(\dots)}$$

$$\rightarrow 0 \text{ as } k \rightarrow \infty$$

$$\Rightarrow \left( \underline{A}_{\hat{n}}^d \right) \underline{u}_0 = v \underline{u}_0$$

Eigenvalue Problem

- if  $\underline{A}_{\hat{n}}^d$  ~~is~~ complete set of e-vectors and all real e-values  
 $\Rightarrow$  full representation of  $\underline{u}(\vec{x}, t)$  as sum of oscillatory, travelling modes  
 $\Rightarrow$  good "strongly hyperbolic"
- conveniently satisfied if  $\underline{A}_{\hat{n}}^d$  is symmetric or symmetrizable  
 $\Rightarrow$  "symmetric hyperbolic"
- for  $\hat{n} \perp \partial\Omega$  (outward), ~~only~~ modes with  $v \geq 0$  leave  $\Omega$   
 $\Rightarrow$  BCs only on modes with  $v < 0$ . "incoming modes"

- if  $\underline{A}^{in}_j$  has real e'values, but no complete set of e'vectors  
"weakly hyperbolic"

Thm: symmetric & strongly hyperbolic systems are well-posed.  
Weakly hyperbolic ones are not.

well-posed  $\cong$   $\exists$  solution

solution is unique

solution depends continuously on initial data

solution is bounded

$$\|u(\cdot, t)\| < e^{Ct} \|u(\cdot, 0)\|$$

Example

$$\begin{aligned} \dot{u}_1 + u_2' &= 0 \\ \dot{u}_2 &= 0 \end{aligned}$$

$$A = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}$$

e'values  $\lambda_1 = \lambda_2 = 0$

only e'vector  $\begin{pmatrix} 0 \\ 1 \end{pmatrix}$

$\Rightarrow$  weakly hyperbolic

take  $u_1(x, 0) = 0, u_2(x, 0) = \epsilon e^{ikx} \Rightarrow \|u(\cdot, 0)\| \sim \epsilon$

but  $\dot{u}_1 = -u_2' = -ikx e^{ikx} \Rightarrow u_1(x, t) = -ikx e^{ikx} t, \|u(\cdot, t)\| \sim \epsilon k t$

arbitrarily fast blow-up  $\uparrow$   
the higher numerical resolution,  
the smaller  $k$  are representable  
the faster the soln blows up  $(\ddot{\smile})$

# NUMERICAL SOLUTIONS

Big Q: 1) How to approximate  $\begin{cases} u(\vec{x}) \\ u(\vec{x}, t) \end{cases}$  by  $\begin{cases} u_h(\vec{x}) \\ u_h(\vec{x}, t) \end{cases}$

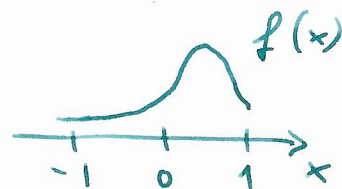
with finite # of DoF?

2) In what sense does  $\begin{cases} u_h(\vec{x}) \\ u_h(\vec{x}, t) \end{cases}$  satisfy the original PDE?

first example: 1-D Laplace Eqn

$$u''(x) = g(x) \quad x \in [-1, 1]$$

ns.t. solution is  $f(x) = e^{-\frac{(x-0.3)^2}{w^2}}$



i.e.  $g(x) = f''(x) = \left( -\frac{2}{w^2} + \frac{4(x-0.3)^2}{w^4} \right) f(x)$

BCs  $u(-1) = f(-1)$

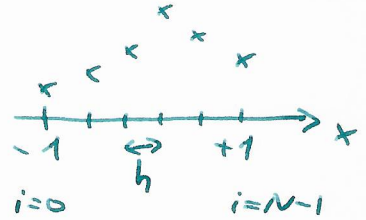
$u(1) = f(1)$

Finite-Differences

N Equally spaced points  $x_i = -1 + h \cdot i \quad h = \frac{2}{N-1}, i = 0, \dots, N-1$

represent solution at  $x_i$

$u(x) \rightarrow u(x_i) \equiv: u_i \quad i = 0, \dots, N-1$



approximate derivs by finite differences

$$\partial_x u(x_i) = \frac{u(x_{i+1}) - u(x_{i-1}))}{2h} = \frac{u_{i+1} - u_{i-1}}{2h} + O(h^2)$$

$$\partial_x^2 u(x_i) = \frac{u_{i+1} - 2u_i + u_{i-1}}{h^2} + O(h^2)$$

$$u'' = f \Rightarrow \frac{u_{i+1} - 2u_i + u_{i-1}}{h^2} = f(x_i) \quad i = 1, 2, \dots, N-2$$

BC  $u_0 = f(-1)$   
 $u_{N-1} = f(+1)$

$$\frac{1}{h^2} \begin{pmatrix} h^2 & & & & & & & \\ & 1 & -2 & 1 & & & & \\ & & & 1 & -2 & 1 & & \\ & & & & & & \ddots & \\ & & & & & & & 1 & -2 & 1 \\ & & & & & & & & & \frac{1}{h^2} \end{pmatrix} \begin{pmatrix} u_0 \\ u_1 \\ u_2 \\ \vdots \\ u_{N-2} \\ u_{N-1} \end{pmatrix} = \begin{pmatrix} f(-1) \\ f_1 \\ f_2 \\ \vdots \\ f_{N-2} \\ f(+1) \end{pmatrix}$$

Matrix Eq  $A \cdot u = b \Rightarrow$  Solve (see ~~Ex~~ Tim)

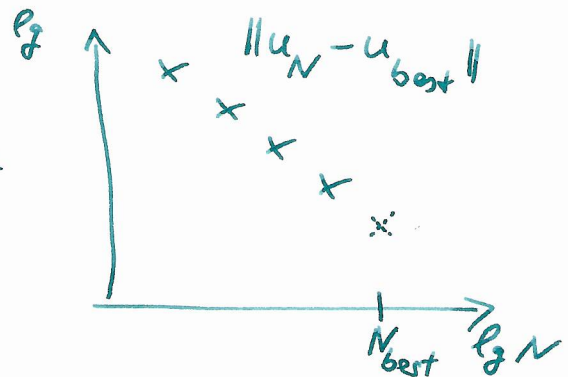
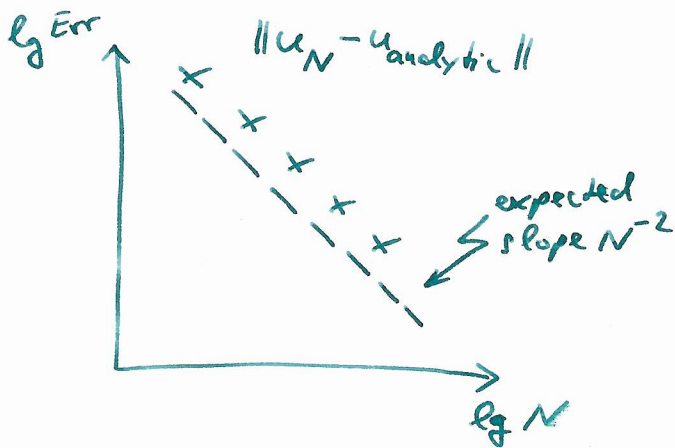
= PDE\_1D\_Poisson.ipynb (part 1) =

IMPORTANT: CHECK CONVERGENCE

COMPARE W/ EXPECTATIONS

→ strong test of correctness

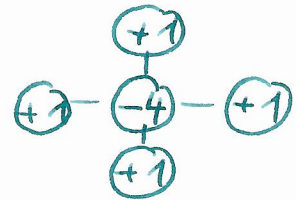
→ tells you how good soln is



2nd order FD:

$$u'' = \frac{u_{i+1} - 2u_i + u_{i-1}}{h^2} + O(h^2) \Rightarrow \text{Err} \propto h^2 \propto N^{-2}$$





Laplace Eq in 2-D with FD

- very similar  $\nabla^2 u_{ij} = \frac{u_{i+1,j} + u_{i-1,j} + u_{i,j+1} + u_{i,j-1} - 4u_{ij}}{h^2}$

- length (b) =  $N^2$  # of grid-points

- <sup>shape</sup> size (A) =  $N^2, N^2$  total  $N^4$  entries!

- indexing more difficult ~~more~~

= PDE\_2D\_Poisson.ipynb =

Curse of dimensionality

$N$  - pts per dim

$N^D$  - # of grid-points

$N^{2D}$  - # of elements in A

$O(N^{3D})$  - ~~cost~~ cost of LU decomp of A.

$N^6$  in 2D

$N^9$  in 3D



# SPECTRAL METHODS

$$\text{approx } u(x,t) \Rightarrow u^{(N)} = \sum_{k=0}^{N-1} \tilde{u}_k(t) \phi_k(x)$$

Fourier  
Chebyshev  
Ylm

smooth solutions  $\Rightarrow$  exponential convergence

$$\|u^{(N)} - u\| \rightarrow e^{-cN} \quad \tilde{u}_k \sim e^{-ck}$$

exact derivatives

$$\partial_x u^{(N)}(x) = \sum_k \tilde{u}_k \underbrace{\partial_x \phi_k(x)}_{\tilde{D}_{ek} \phi_e(x)} = \sum \tilde{u}'_e \phi_e$$

$\uparrow \tilde{D}_{ek} \tilde{u}_k$

physical  $\leftrightarrow$  spectral

$x_i$   $i=0, \dots, N-1$  non-uniform

$$u_i := u(x_i) = \sum \tilde{u}_k \underbrace{\phi_k(x_i)}_{\equiv A_{ik}} = A_{ik} \tilde{u}_k$$

also  $\tilde{u}_k = (A^{-1})_{ki} u_i$

good choice  $\{x_i\} \leftrightarrow$  well-conditioned  $A_{ik}$   
& trafo doable by FFT

Boyd: "Chebyshev and Fourier Spectral Methods"

Demo: Laplace eq in 1D with Chebyshev

$$T_k(x) = \cos(k \arccos x) \quad x \in [-1, 1]$$

$$x_i = \cos \frac{\pi i}{N-1} \quad i=0, \dots, N-1$$



$$A_{ik} = T_k(x_i)$$

$$A_{kj}^{-1} = \frac{2}{(N-1) \bar{c}_k \bar{c}_j} T_k(x_j) \quad \bar{c}_k = \begin{cases} 2 & k=0, N-1 \\ 1 & \text{else} \end{cases}$$

following  
Kiddler +, PRD 62  
p. 084026 (2008)

Discrete orthogonality

$$\sum_i \frac{2}{(N-1) \bar{c}_k \bar{c}_j} T_k(x_i) T_j(x_i) = \delta_{kj}$$

Differentiation by recurrence formulae

$$u = \sum \tilde{u}_k T_k, \quad u' = \sum \tilde{u}'_k T_k$$

$$\tilde{u}'_k = \frac{1}{c_k} (\tilde{u}'_{k+2} + 2(k+1) \tilde{u}_{k+1}) \quad c_k = \begin{cases} 2 & k=0 \\ 1 & \text{else} \end{cases}$$

$$\text{seed with } \tilde{u}'_N = 0 \quad \tilde{u}'_{N-1} = 0$$

↑ deriv of polynomial reduces order

$$\Rightarrow \tilde{u}'_k = \tilde{D}_{ke} \tilde{u}_e$$

Deriv at grid-points

$$u'(x_i) =: u'_i = A_{ik} \tilde{D}_{ke} A_{ej}^{-1} u_j$$

$$u'' = s \iff (A \tilde{D}^2 A^{-1})_{ji} u_i = s_j(x_j)$$

↳ modify by BC  $\Rightarrow$  Matrix Eq again